## ML assignment:

#### Dataset:

Annotated dataset of around 3k images which has locations of defects (bounding box)

#### **Problem statement:**

- Based on the given dataset, identifying if a given image is healthy/ not binary classification
  - Problem statement 1( using custom model from scratch )
- Based on the given dataset, predict multi-label classification using custom model Dataset is not being loaded due to out of memory error ( since it has shape len \* 6 (classes) \* 224 \* 224 \* 3)
- Based on the given dataset , predict & localize multiclasses using pretrained model (By fine tuning the model, making it customized to the dataset)
  - Problem statement 2

Classification folder refers to problem statement 1 Detection folder refers to problem statement 2

#### **Problem statement 1:**

### Pipeline:

- Data Preparation (loading the data & modifying the paths as per requirement)
- Data preprocessing
  resize the image (since we have multiple shapes) to (224,224)
  Normalize the pixel values to [0,1]
- Data augmentation

**Flipping** 

Rotation

Brightness

Width, height shift

- Models
  - Manual feature selection:

Trained the model using HOG, LBP and use random forest for classification

- CNN model: (auto feature selection using CNN & trained with augmented dataset)

### - Inferences

#### **Problem statement 2:**

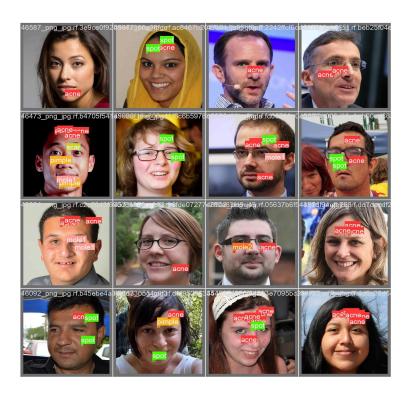
## Pipeline:

- Extracted the dataset from roboflow (in yolov5 format)
- Seperated the dataset into train, validation classes for gaining the prediction
- Adjusted the path variables in .yml file & constrained hyperparameters for low training time.

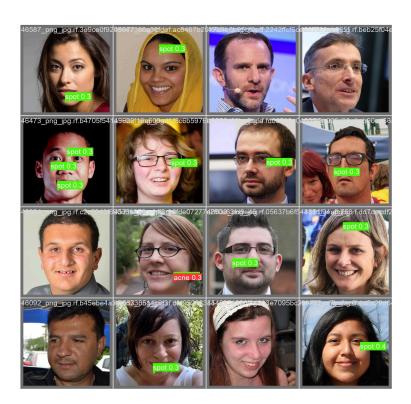
#### Inferences:

- Following defects are identified in a better way in the given order: *spot*, *acne*, *mole*, *pimple*..

# Actual image:



# Predicted:



Predictions couldn't identify acne/pimple/ other classes in better way compared to spots due to:

- low epochs: 5
- Since the model is taking very high training time: around 37 minutes for a epoch, I've restricted the model to 5 epochs instead of general 20/50 range
- Low quality dataset:
   Dataset isn't preprocessed (resized ) & augmented (which reduces its robustness)
   Also the dataset lacks the diversity among the classes & annotation isn't upto the mark for some images as well
  - Too small bounding box: Adding to the less amount of data, bounding box size is also too small which is making it difficult to differentiate between similar classes

#### **Future work:**

- Preprocess the dataset before hand to ensure all images are of similar shape
- Normalize the image pixels as part of pre-processing for better convergence
- Augmented the dataset by flipping, rotating, brightness, contrast ... to ensure robustness
- Increase the epochs and ensure model is being trained with GPU for faster execution
- If model predicts similar to the labels, we can evaluate using IOU & other evaluation metrics
- Mole class can be removed since it doesn't come under curable defect, instead we can add whiteheads ( manual annotation ...)